



Liquid Argon Time Project Chamber In a testbeam

The little detector that could.



particle classification using Neural networks

Samuel Mathias Borer
The University of Maine - Orono

Liquid Argon in a testbeam

The little detector with a big heart.

“A billion
Neutrinos go
swimming: one
gets wet.”
michael
kamakana

Deep
Learning

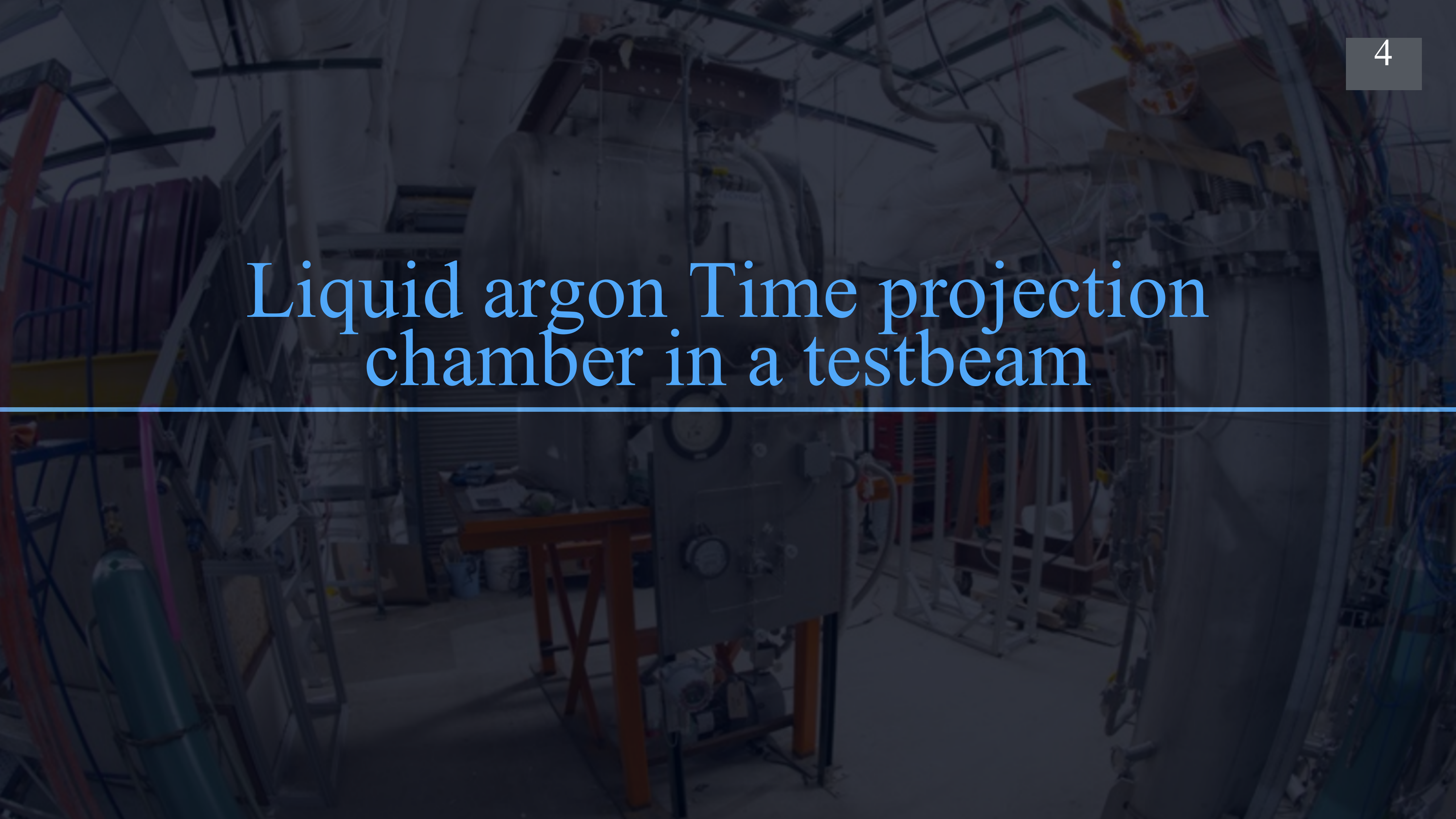
01

02

current particle
identification
methods

03

Liquid argon Time projection chamber in a testbeam



Why liquid argon?

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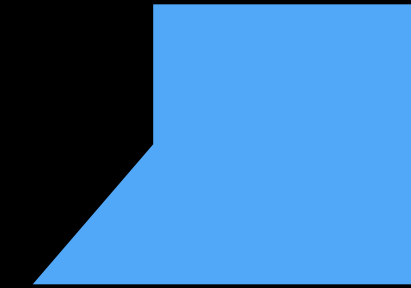
Abundance

Argon makes up 1% of the atmosphere



Density

Liquid argon is 40% more dense than water.



Ionization

Argon can be readily ionized.



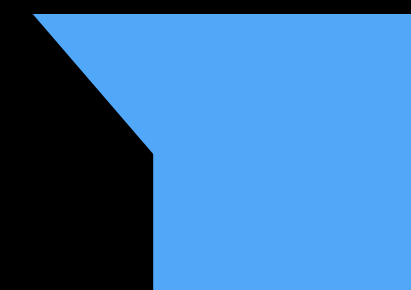
Scintillation

Produces light and is transparent to the light produced.

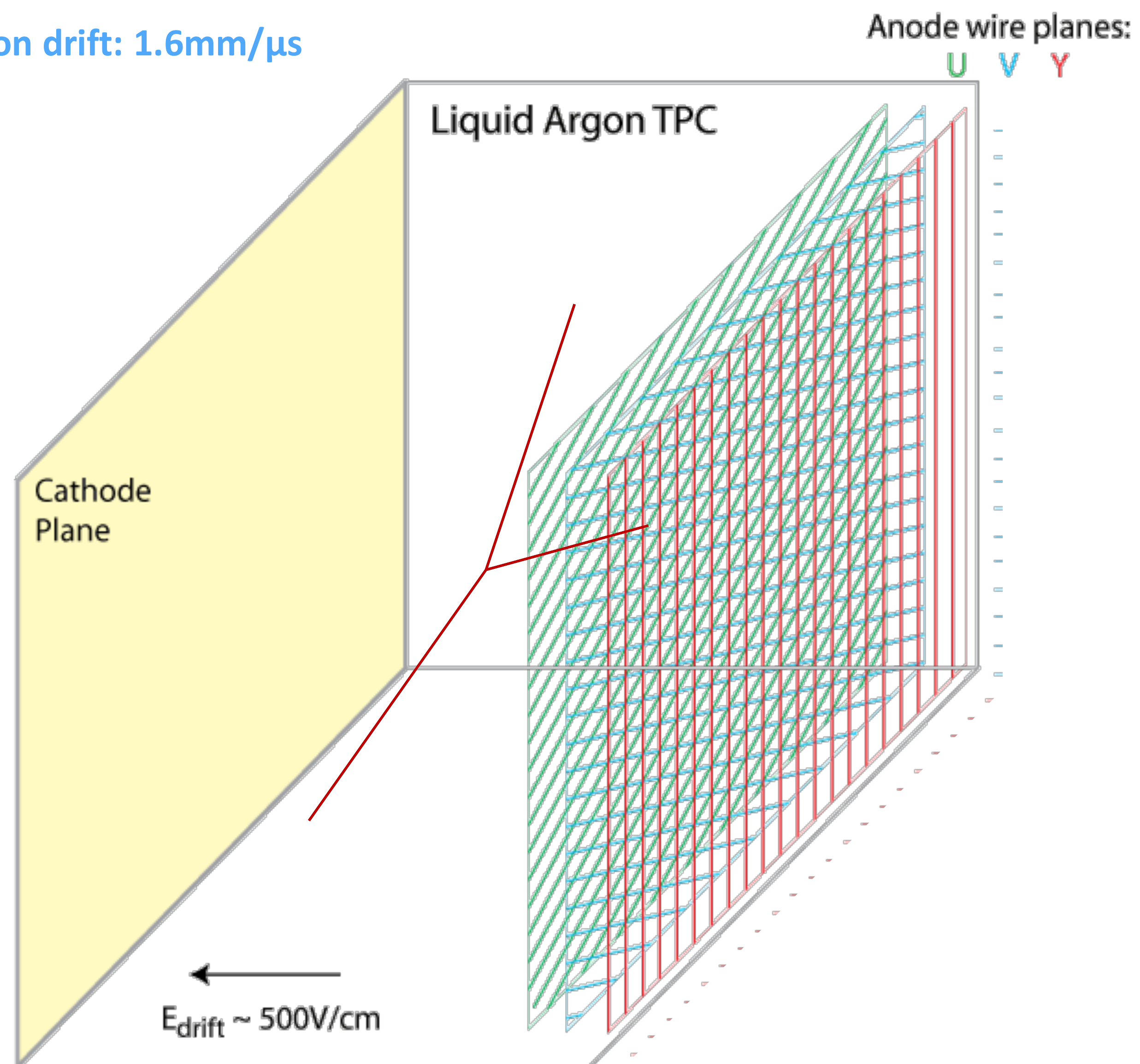


Electron Lifetime

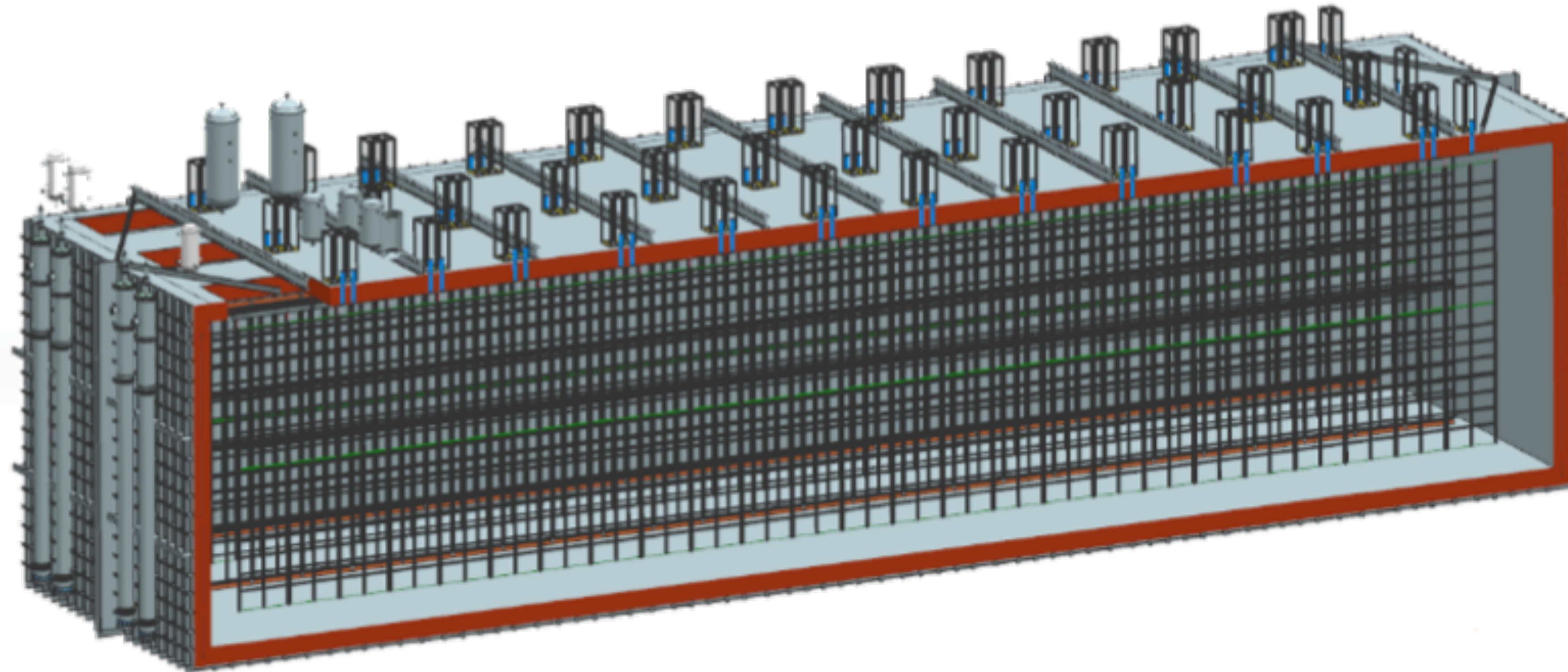
Greek name means “inactive”



Electron drift: $1.6\text{mm}/\mu\text{s}$

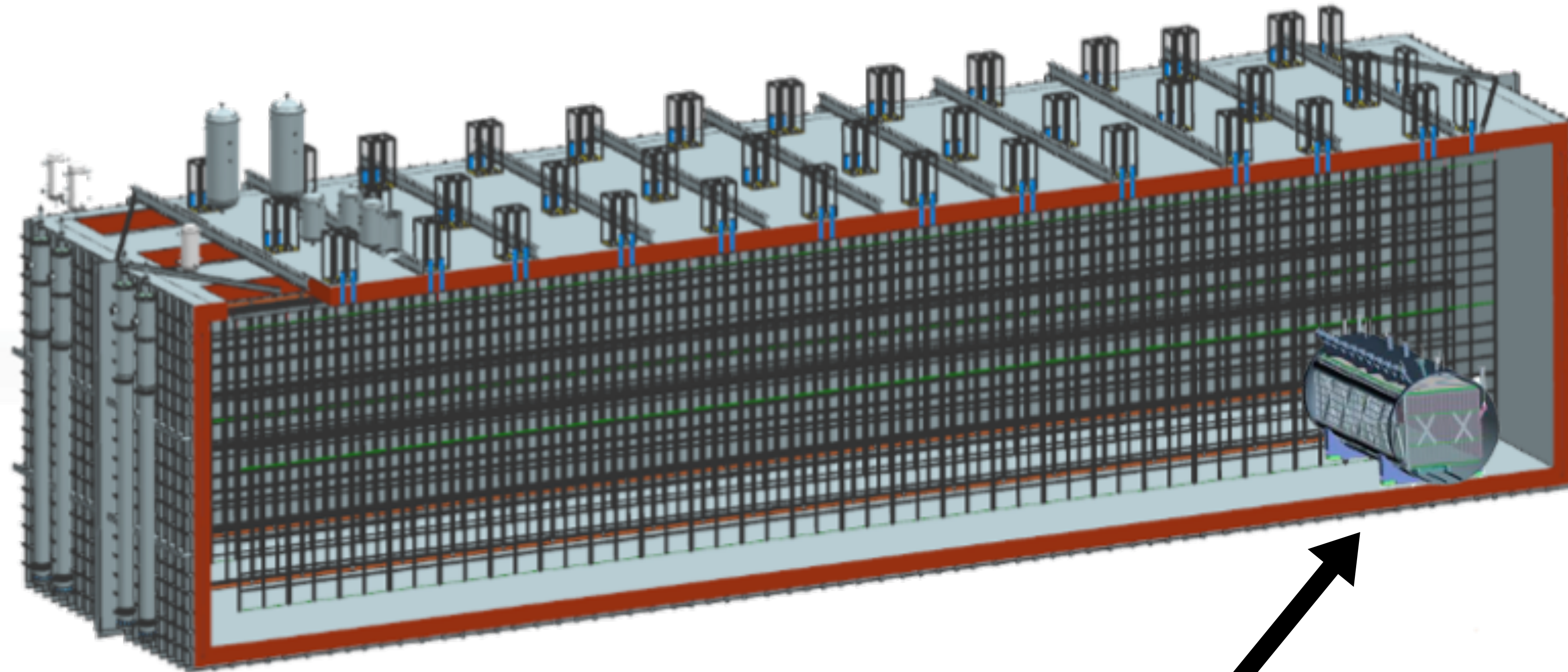


Liquid Argon Time Projection Chamber



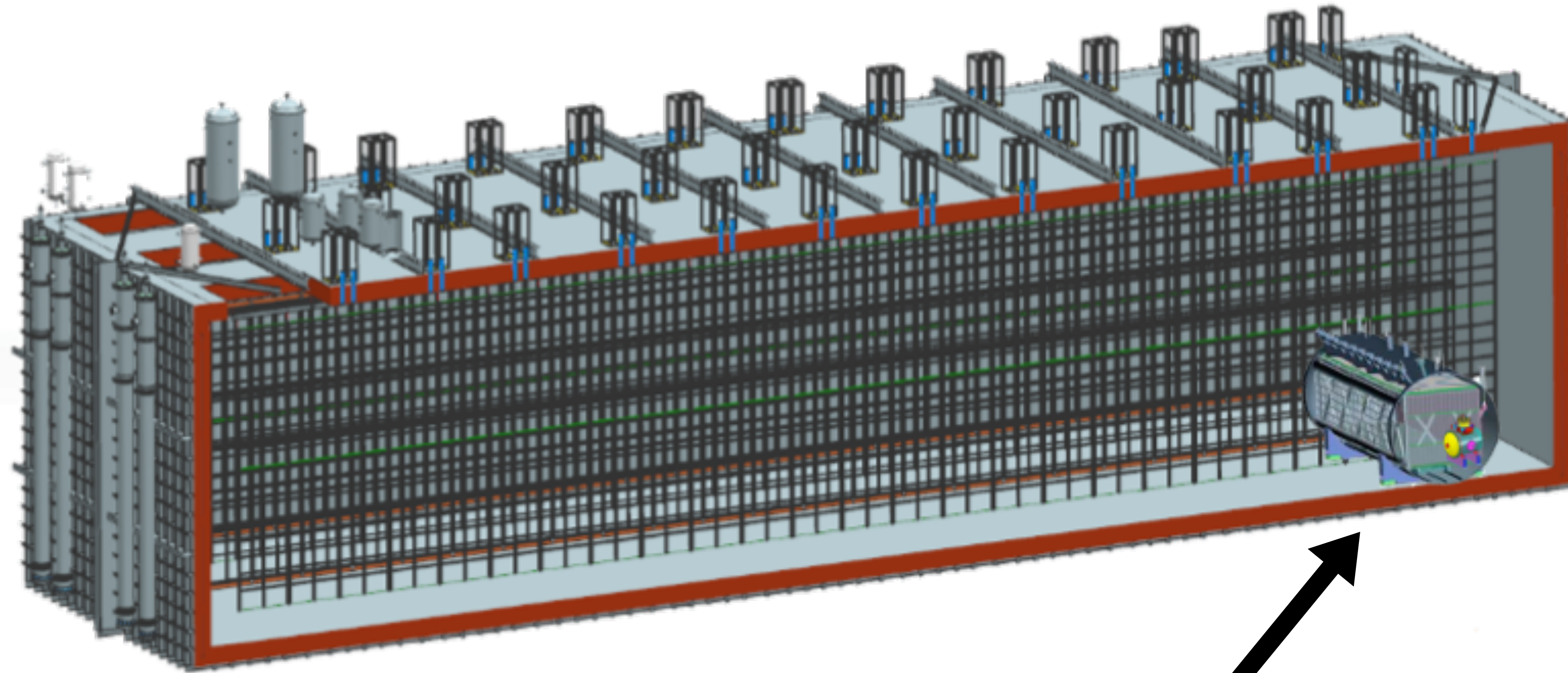
Liquid
Argon
Time
Projection
Chamber

One 10kT DUNE LArTPC Module
18m x 19m x 66m



Liquid
Argon
Time
Projection
Chamber

MicroBooNE TPC (80 tons)
2.2m x 2.5m x 10m

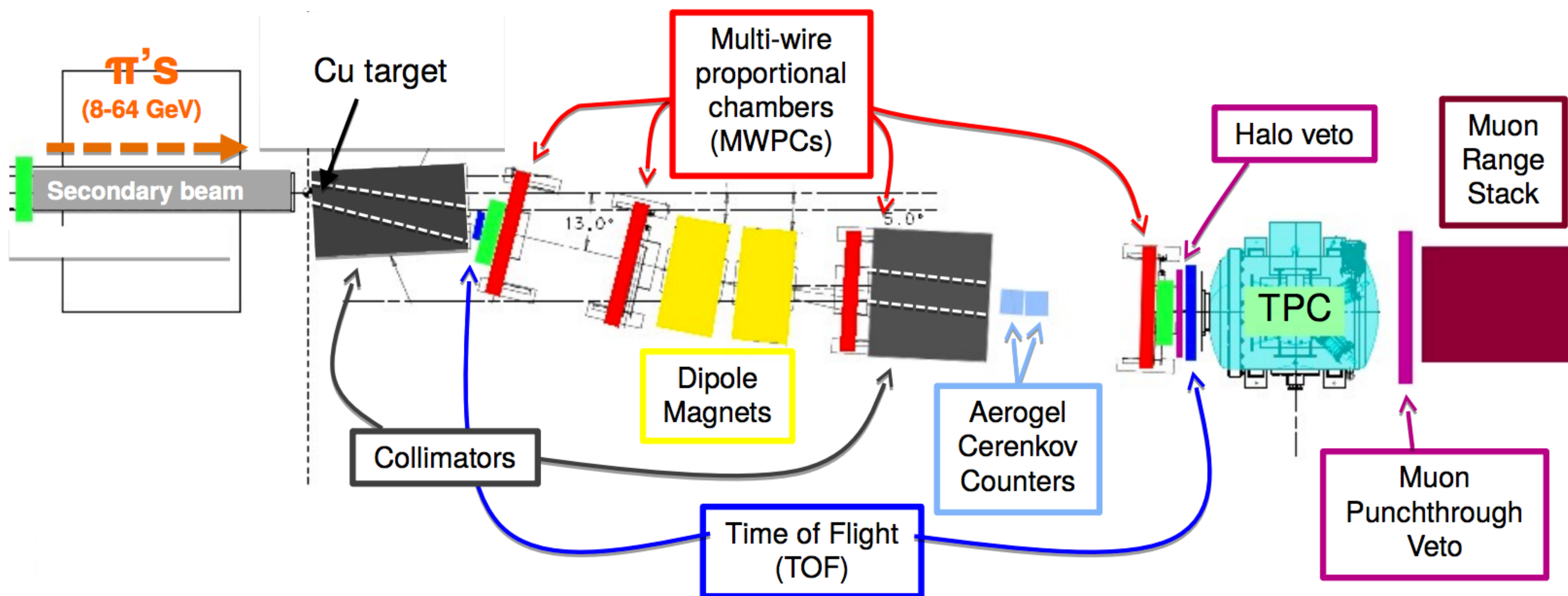


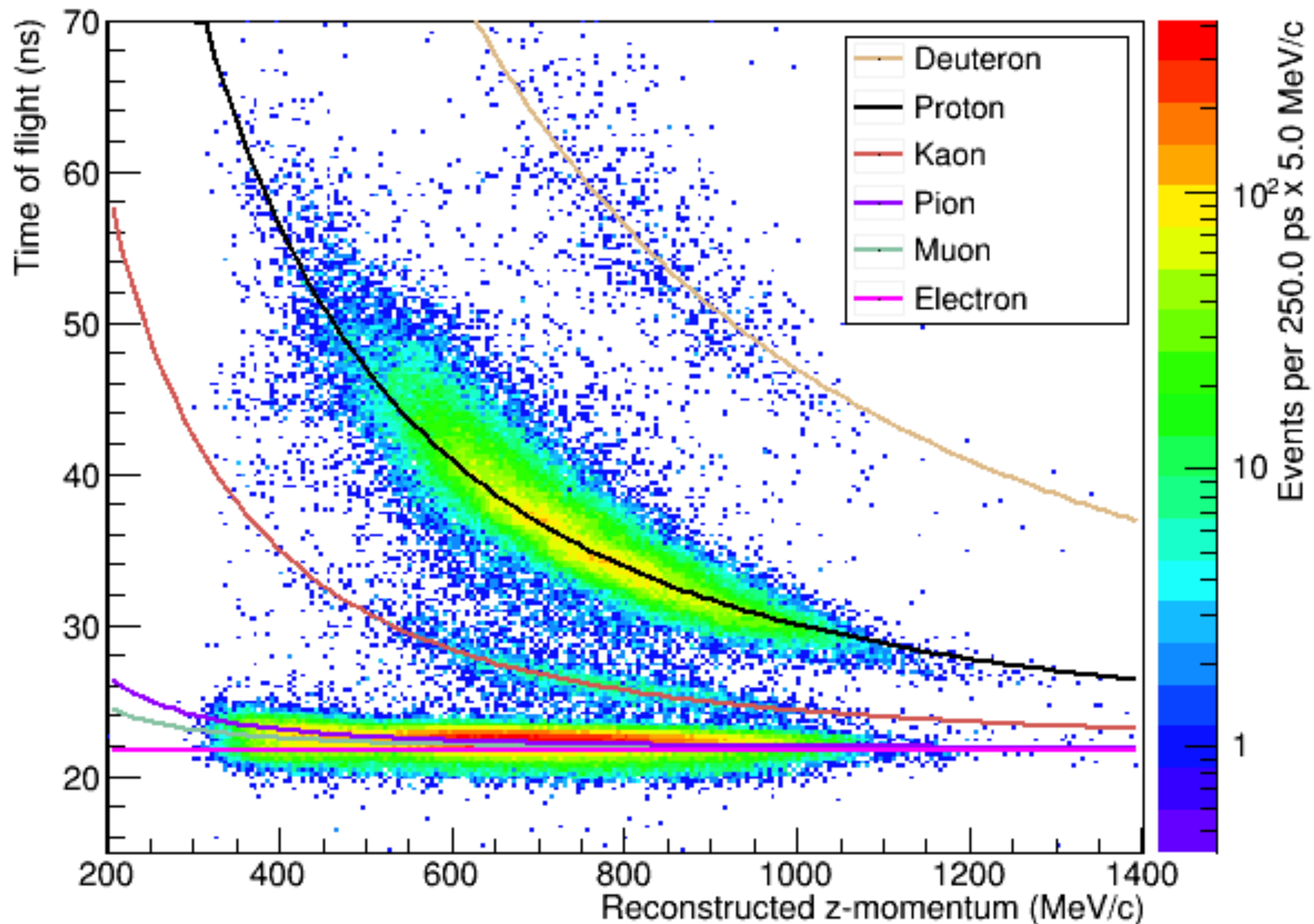
LArIAT TPC (0.25 tons)
0.4m x 0.47m x 0.9m

Liquid
Argon
Time
Projection
Chamber

Current particle identification methods

Liquid Argon In a testbeam





Liquid
Argon
In
a
testbeam

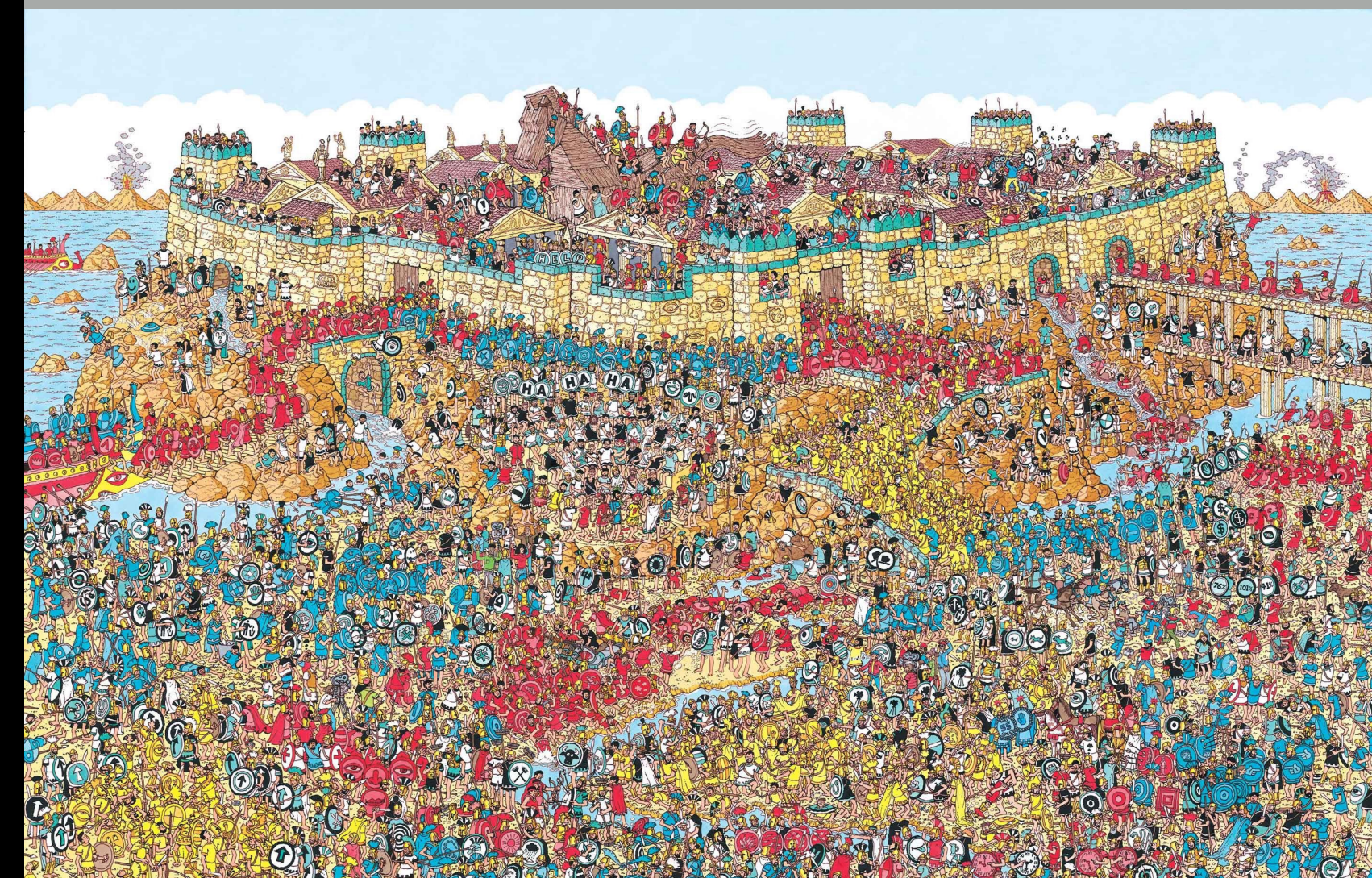
Deep Learning

A A A A A

Template Matching



Prototype Matching



Feature Analysis

Pattern Recognition

Two kinds of machine learning

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Supervised Learning

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output relationship

In supervised learning, we are given a data set and already know what the correct output should look like.

Regression

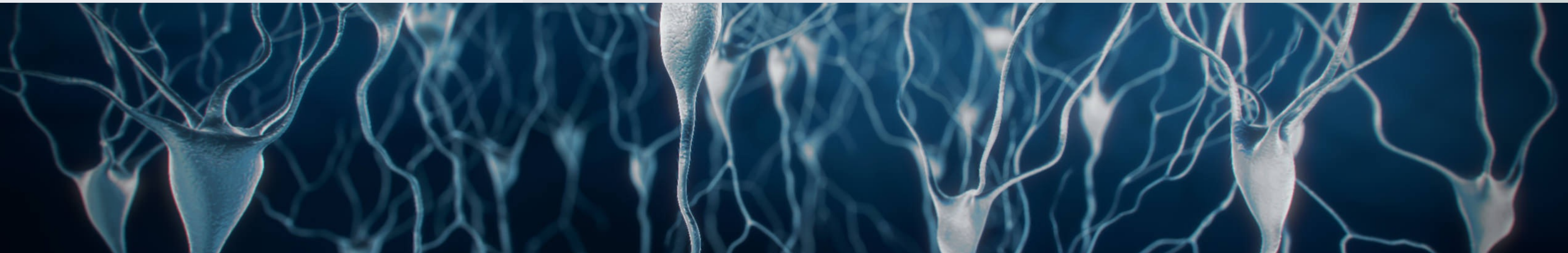
In regression problems, we are trying to predict results within a continuous output.

Example: An output trying to predict the price of houses based on the real estate market

Classification

In classification problems, we are trying to predict results with a discrete output.

Example: An output to determine whether a house will sell for more or less than the asking price.



UnSupervised Learning

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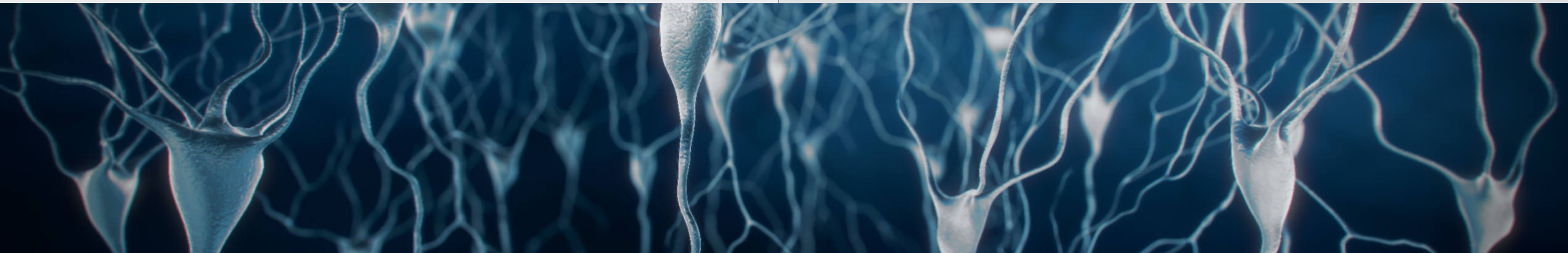
No desired output

In unsupervised learning, we can approach problems with no particular idea of what our results will look like.

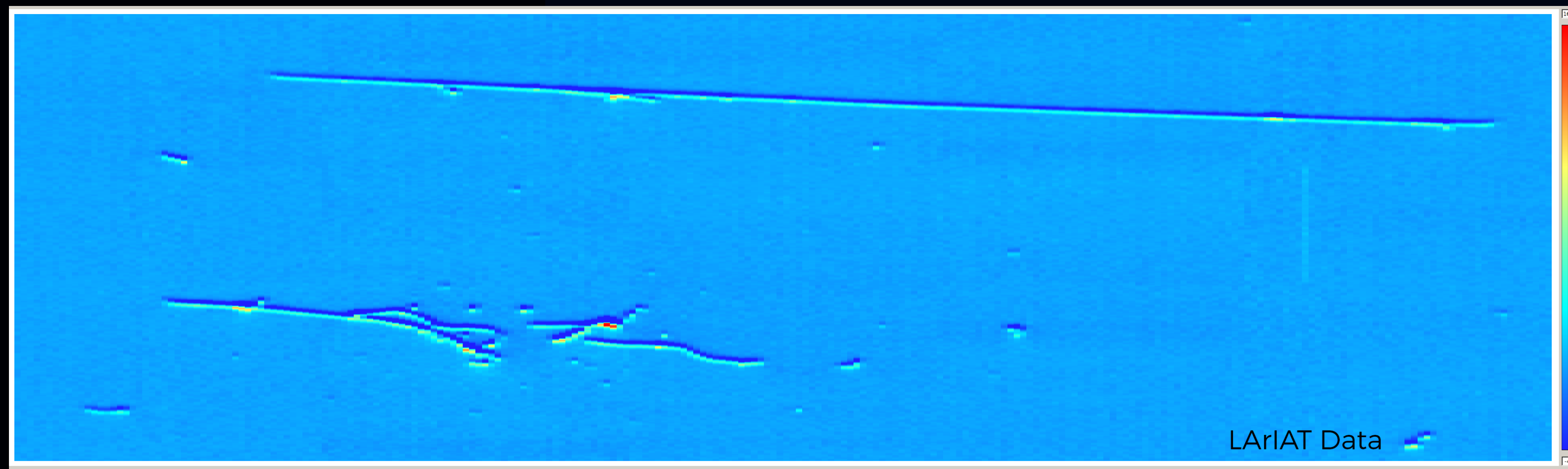
Our structure will cluster data based on undefined variables to find patterns and relationships.

clustering

Say you were handed a stack of note cards with pictures and words on them. You could feed them into an unsupervised learning software and it will try to find a way to group these notecards based on features it sees.



I want to be able to feed our event display images into a neural network who will then classify the tracks into specific particles.



Our
Problem



Convolutional Neural Network

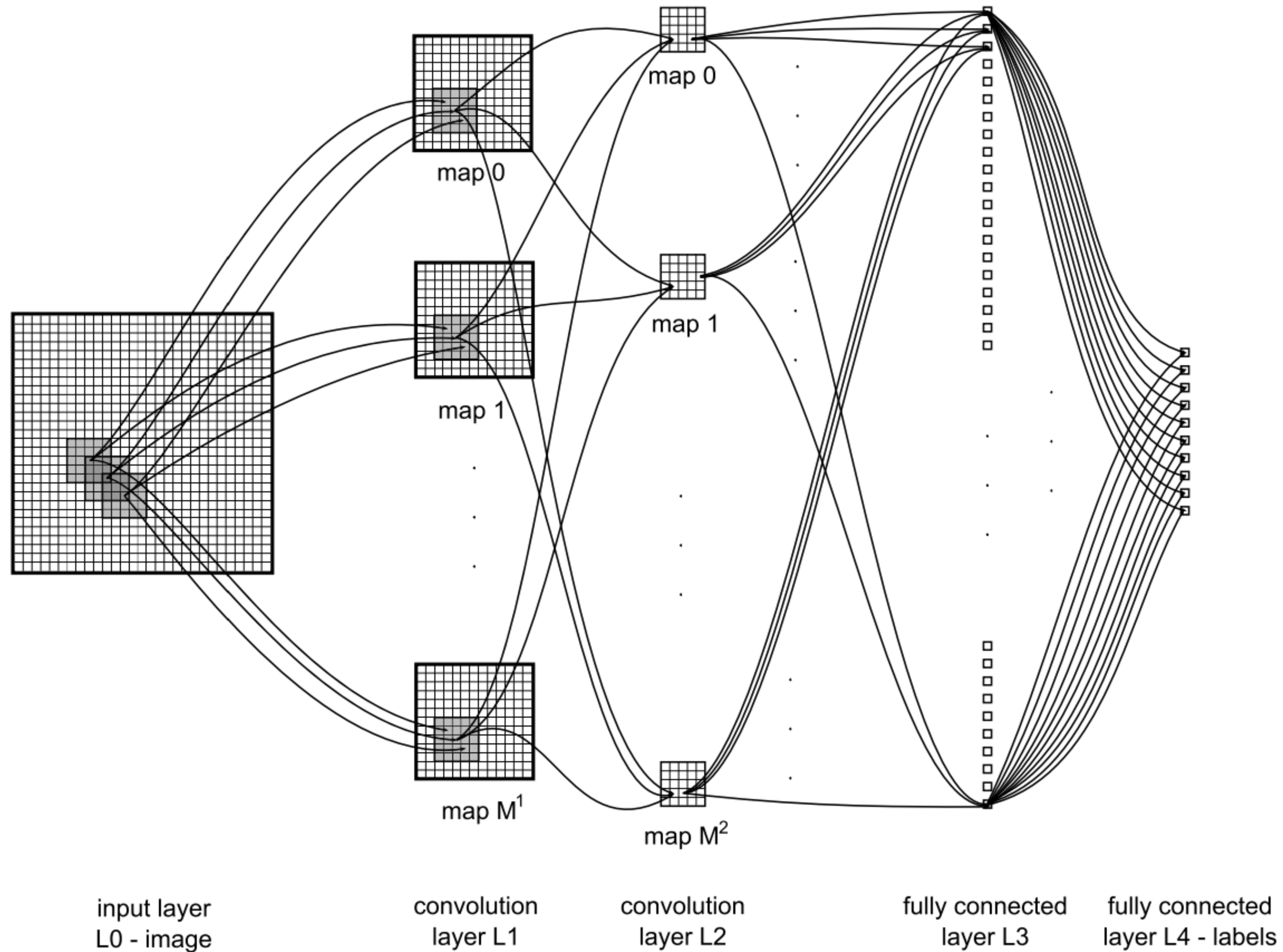
Convolutional Neural Networks are based off of biology. A study conducted at Harvard Medical School on macaque and spider monkeys found three classifications of the cells of the visual cortex.



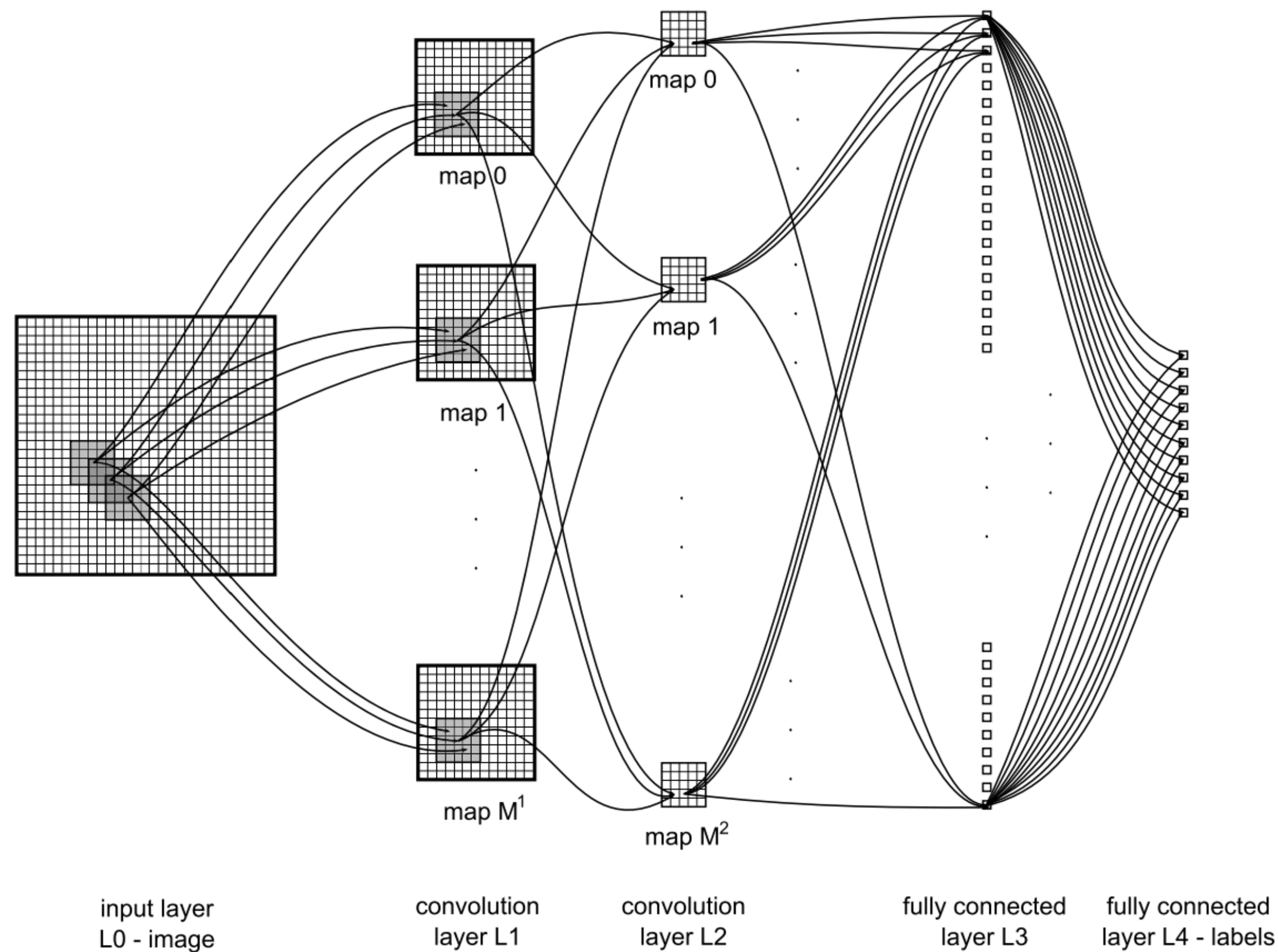
Simple Cells



Complex & Hypercomplex



Receptive Fields

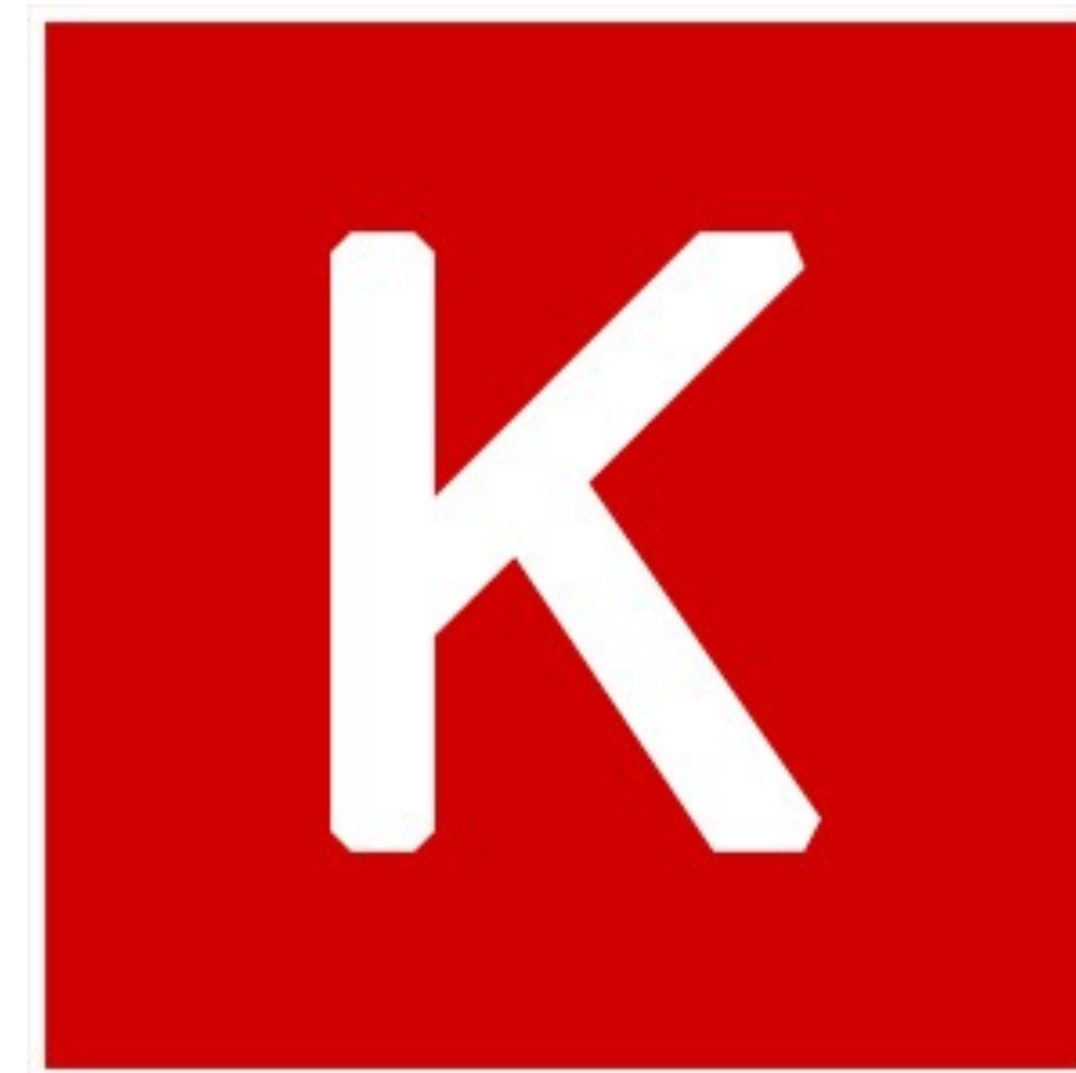


Shared weights

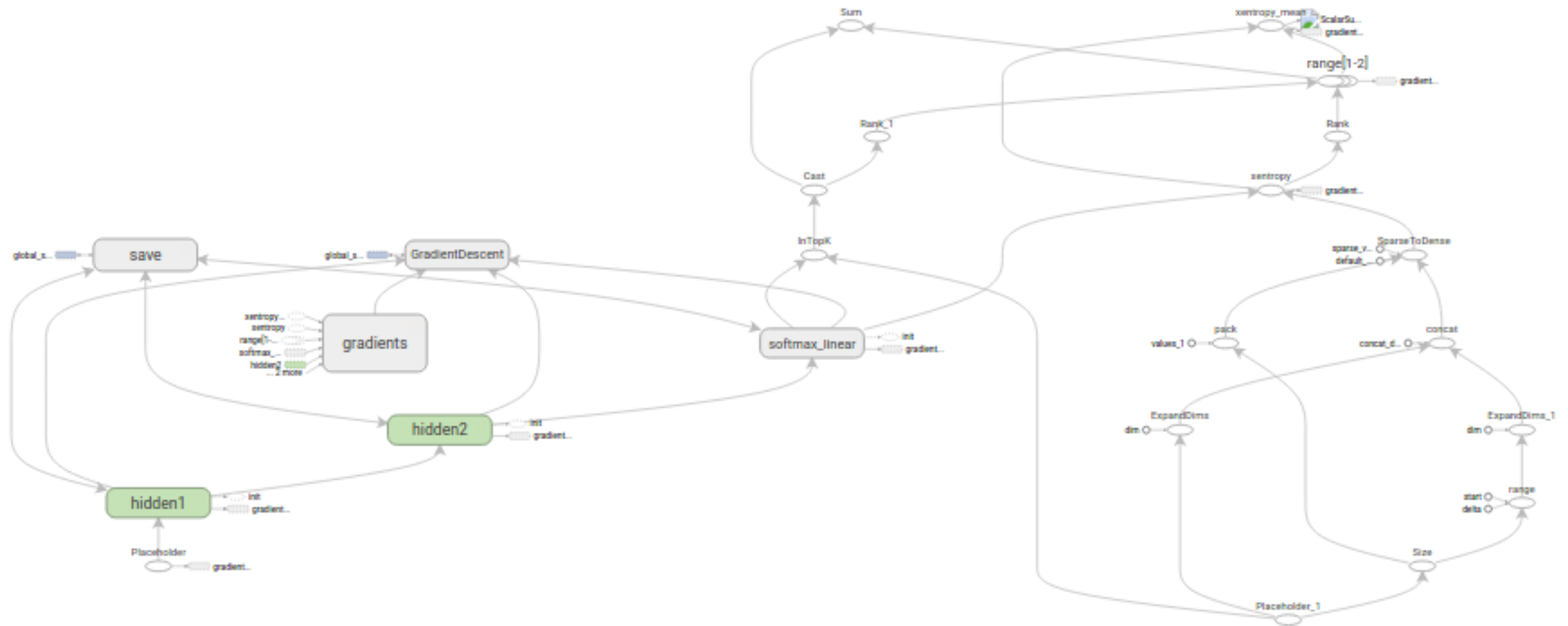
Each input is modified by a *weight*, which is the relative amplitude of the connection between two nodes and it defines the amount of influence one node will have on the other.

Tensor flow & keras

We have designed a way to take reconstructed TPC images and convert them to an input for TensorFlow. We are currently in the process of designing the multi-layered structure of the CNN to test the first run of training.

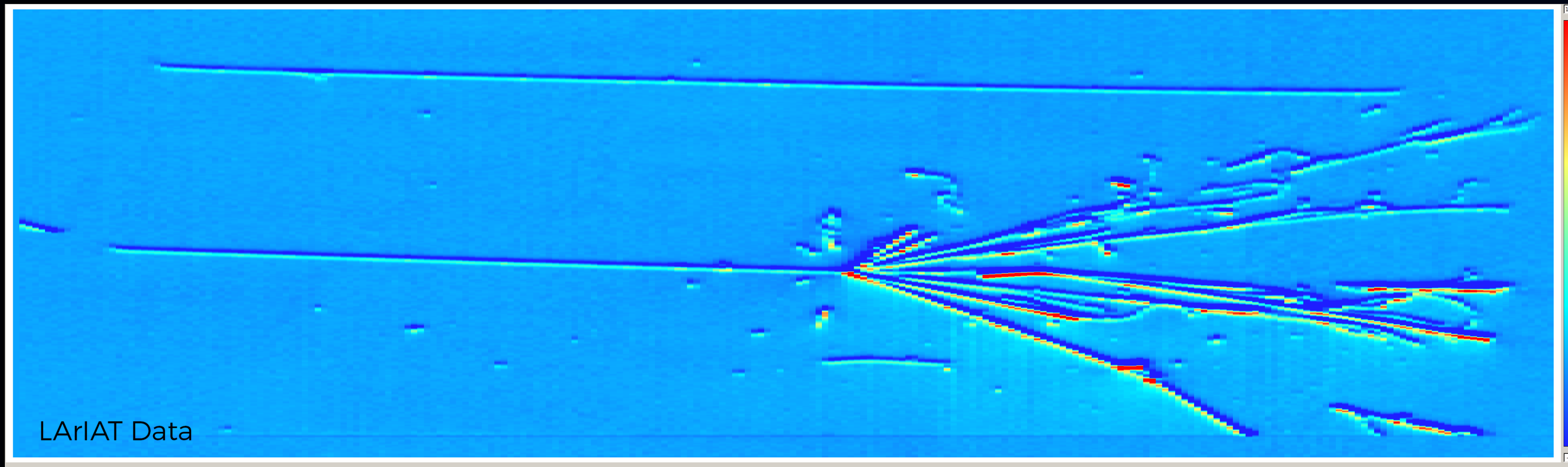


Tensorflow & keras



Currently, we are working to build the framework of our Convolutional Neural Network and use training sets we have created from our data to teach it to classify each particle interaction. By using LArIAT data with known inputs and outputs we can then prepare our CNN for future liquid argon neutrino experiments such as DUNE.

Next
Steps



To the LArIAT
Collaboration,
Dr. Jennifer Raaf,
Dr. Jason St. John,
Dr. Roberto Acciarri,
Gregory Pulliam,
Daniel Smith,
Derek Walker,
and the SIST Committee.



thank you!